```
Neural Network with all categories
        import xarray as xr
In [1]:
         import numpy as np
         import torch
         from torch.utils.data import Dataset
        from torch.utils.data import DataLoader
         import torch
         from torch import nn
In [2]: class TemperatureLandcover1D(Dataset):
            def init (self, path, subset="train", seed=87, normalize=False, reduce
                 self.path = path
                 self.subset = subset
                 self.seed = seed
                 # Load the data
                 ds = xr.open_dataset(self.path)
                 t2m = ds.air_temperature_at_2_metres.values # (143, 489, 529)
                 t30m = ds.air_temperature_at_30_metres.values # (143, 489, 529)
                 landcover = ds.landcover.values # (member, x, y) int8
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# To include all categories, get the landcover categories
landcover_classes = np.unique(landcover)
# Create a dictionary to map landcover categories to integers
landcover_to_ix = {category: i for i, category in enumerate(landcover_to_ix = {category: i for i, category in enumerate(landcover_to_ix = {category: i for i, category: i for i for i, category: i for i 
# Flatten the data
t2m = t2m.flatten()
t30m = t30m.flatten()
# Replace the landcover variable categories to an array of integers
landcover = np.array([landcover_to_ix[category] for category in land
# Normalize the temperature data
if normalize:
           t2m = (t2m - t2m.mean()) / t2m.std()
           t30m = (t30m - t30m.mean()) / t30m.std()
# Split the data into training, testing, and validation subsets
np.random.seed(self.seed)
if reduce_to is None:
            reduce_to = t2m.size
train_idxs = np.random.randint(0, t2m.size, reduce_to // 2) # 50%
test_idxs = np.random.randint(0, t2m.size, reduce_to // 4) # 25% i
val_idxs = np.random.randint(0, t2m.size, reduce_to // 4) # 25% in
if self.subset == "train":
            idxs = train_idxs
elif self.subset == "test":
            idxs = test_idxs
elif self.subset == "val":
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idxs = val idxs
                 else:
                     raise ValueError("Unknown subset")
                 # Convert the selected indices to tensors
                 # ensure idxs variable is within the bounds of the landcover array L
                 # modulus operator (%) to ensure that the index is within the range
                 self.t2m = torch.tensor(t2m[idxs % t2m.size])
                 self.t30m = torch.tensor(t30m[idxs % t30m.size])
                 self.landcover = torch.tensor(landcover[idxs % landcover.size])
            def __len__(self):
                 return len(self.t2m)
                  _getitem__(self, idx):
                 """Return an item of the dataset.
                 Parameters
                 idx: int
                     Index of the item to return
                 Returns
                 x: float tensor of shape (1,)
                    Temperature at 2m
                 c: int64
                     Landcover category
                 y: float tensor of shape (1,)
                    Temperature at 30m
                 x = self.t30m[idx].unsqueeze(0)
                 c = self.landcover[idx].long()
                 y = self.t2m[idx].unsqueeze(0)
                 return x, c, y
In [ ]:
In [3]: ds = TemperatureLandcover1D("/Users/marwa/Desktop/MyWorkingdirectory/interns
        x, c, y = ds[0]
        tensor([280.3035])
Out[3]:
In [4]: class LinearRegressionWithEmbedding(nn.Module):
            def __init__(self, n_landcovers=33, embedding_size=3, output_size=1):
                 super().__init__()
                 self.landcover_embedding = nn.Embedding(n_landcovers, embedding_size
                 self.linear = nn.Linear(embedding_size + 1, output_size)
            def forward(self, x, c):
                 emb = self.landcover_embedding(c)
                 x = torch.cat([x, emb], dim=1)
                 return self.linear(x)
In [ ]:
In [5]:
        # Load dataset
        # __
        #specify batch size
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batch size = 4096
#specify file path
ncfile = "/Users/marwa/Desktop/MyWorkingdirectory/internship-marwa/data/mera
#The TemperatureLandcover1D class is used to create instances of the dataset
#The subset parameter is set to "train" for the training data and "test" for
#The normalize parameter is set to True for both datasets.
#The reduce_to parameter is set to 3,000,000 for both datasets.
training_data = TemperatureLandcover1D(
    ncfile, subset="train", normalize=True, reduce_to=3_000_000
test_data = TemperatureLandcover1D(
    ncfile, subset="test", normalize=True, reduce to=3 000 000
x0, c0, y0 = training_data[0]
print(f"The dataset has {len(training_data)} items. Each item looks like {x@
train_dataloader = DataLoader(training_data, batch_size=batch_size)
test dataloader = DataLoader(test data, batch size=batch size)
x, c, y = next(iter(train dataloader))
print(f"The dataloader create batches of items of shape {x.shape, c.shape, y
# Network
# -----
net = LinearRegressionWithEmbedding()
y_pred = net(x, c)
# Loss and optimizer
loss_fn = torch.nn.MSELoss()
print(f"Loss value: {loss_fn(y_pred, y)}")
optimizer = torch.optim.SGD(net.parameters(), lr=1e-3)
# Training
# --
device = "cuda" if torch.cuda.is_available() else "cpu"
print(f"Using {device} device")
def train(dataloader, model, loss_fn, optimizer):
    size = len(dataloader)
    model.train()
    for batch, (x, c, y) in enumerate(dataloader):
        x, c, y = x.to(device), c.to(device), y.to(device)
        # Compute prediction error
        pred = model(x, c)
        loss = loss_fn(pred, y)
        # Backpropagation
        loss.backward()
        optimizer.step()
        optimizer.zero_grad()
        if batch % 100 == 0:
```

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loss, current = loss.item(), batch
                                                print(f"loss: {loss:>7f} [{current:>5d}/{size:>5d}]")
                   def test(dataloader, model, loss_fn):
                             num batches = len(dataloader)
                             model.eval()
                             test loss = 0
                             baseline = 0
                             with torch.no grad():
                                      for x, c, y in dataloader:
                                                x, c, y = x.to(device), c.to(device), y.to(device)
                                                pred = model(x, c)
                                                test_loss += loss_fn(pred, y).item()
                                                baseline += loss fn(x, y).item()
                             test_loss /= num_batches
                             print(
                                       f"Test Error: \n Mean squared error: {test_loss:>8f} (baseline: {baseline: {b
                    epochs = 2
                    for t in range(epochs):
                             print(f"\nEpoch {t+1}\n----
                             train(train_dataloader, net, loss_fn, optimizer)
                             test(test dataloader, net, loss fn)
                    print("Done!")
                   The dataset has 1500000 items. Each item looks like (tensor([-0.6996]), ten
                   sor(0), tensor([-0.5873]))
                   The dataloader create batches of items of shape (torch.Size([4096, 1]), tor
                   ch.Size([4096]), torch.Size([4096, 1]))
                   Loss value: 0.7830725908279419
                   Using cpu device
                   Epoch 1
                   loss: 0.783073
                                                                              367]
                                                                    0/
                   loss: 0.471612
                                                               100/
                                                                              367]
                   loss: 0.284469
                                                      [
                                                               200/
                                                                              367]
                   loss: 0.185182
                                                       [ 300/
                                                                              367]
                   Test Error:
                     Mean squared error: 0.141009 (baseline: 2.821028)
                   Epoch 2
                   loss: 0.139461 [
                                                                    0/
                                                                              367]
                   loss: 0.103497
                                                               100/
                                                                              367]
                   loss: 0.070247
                                                               200/
                                                                              367]
                   loss: 0.053444
                                                       [ 300/
                                                                             367]
                   Test Error:
                     Mean squared error: 0.045604 (baseline: 2.821028)
                   Done!
In [ ]:
                 class LinearRegressionWithEmbedding(nn.Module):
In [6]:
                             def __init__(self, n_landcovers=33, embedding_size=3, output_size=1):
                                      super().__init__()
                                       self.landcover_embedding = nn.Embedding(n_landcovers, embedding_size
                                      self.fc1 = nn.Linear(embedding_size+1, 128)
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self.fc2 = nn.Linear(128, output_size)

def forward(self, x, c):
    emb = self.landcover_embedding(c)
    x = torch.cat([x, emb], dim=1)
    x = torch.relu(self.fc1(x))
    return self.fc2(x)
```

```
In [7]: # Load dataset
        # --
        #specify batch size
        batch_size = 4096
        #specify file path
        ncfile = "/Users/marwa/Desktop/MyWorkingdirectory/internship-marwa/data/mera
        #The TemperatureLandcover1D class is used to create instances of the dataset
        #The subset parameter is set to "train" for the training data and "test" for
        #The normalize parameter is set to True for both datasets.
        #The reduce_to parameter is set to 3,000,000 for both datasets.
        training_data = TemperatureLandcover1D(
            ncfile, subset="train", normalize=True, reduce_to=3_000_000
        test_data = TemperatureLandcover1D(
            ncfile, subset="test", normalize=True, reduce_to=3_000_000
        x0, c0, y0 = training data[0]
        print(f"The dataset has {len(training_data)} items. Each item looks like {x(
        train_dataloader = DataLoader(training_data, batch_size=batch_size)
        test_dataloader = DataLoader(test_data, batch_size=batch_size)
        x, c, y = next(iter(train_dataloader))
        print(f"The dataloader create batches of items of shape {x.shape, c.shape, y
        # Network
        # ---
        net = LinearRegressionWithEmbedding()
        y_pred = net(x, c)
        # Loss and optimizer
        # --
        loss_fn = torch.nn.MSELoss()
        print(f"Loss value: {loss_fn(y_pred, y)}")
        optimizer = torch.optim.SGD(net.parameters(), lr=1e-3)
        # Training
        device = "cuda" if torch.cuda.is_available() else "cpu"
        print(f"Using {device} device")
        def train(dataloader, model, loss_fn, optimizer):
            size = len(dataloader)
            model.train()
            for batch, (x, c, y) in enumerate(dataloader):
                x, c, y = x.to(device), c.to(device), y.to(device)
```

```
# Compute prediction error
                           pred = model(x, c)
                           loss = loss_fn(pred, y)
                           # Backpropagation
                           loss.backward()
                           optimizer.step()
                           optimizer.zero_grad()
                           if batch % 100 == 0:
                                         loss, current = loss.item(), batch
                                         print(f"loss: {loss:>7f} [{current:>5d}/{size:>5d}]")
def test(dataloader, model, loss_fn):
             num_batches = len(dataloader)
             model.eval()
             test_loss = 0
             baseline = 0
             with torch.no_grad():
                           for x, c, y in dataloader:
                                         x, c, y = x.to(device), c.to(device), y.to(device)
                                         pred = model(x, c)
                                         test_loss += loss_fn(pred, y).item()
                                         baseline += loss_fn(x, y).item()
             test_loss /= num_batches
             print(
                           f"Test Error: \n Mean squared error: {test_loss:>8f} (baseline: {baseline: {b
epochs = 2
for t in range(epochs):
             print(f"\nEpoch {t+1}\n------
             train(train_dataloader, net, loss_fn, optimizer)
             test(test_dataloader, net, loss_fn)
print("Done!")
```

The dataset has 1500000 items. Each item looks like (tensor([-0.6996]), tensor([-0.5873]))

The dataloader create batches of items of shape (torch.Size([4096, 1]), torch.Size([4096]), torch.Size([4096, 1]))

Loss value: 1.0053765773773193

Using cpu device

Epoch 1

loss: 1.005377 [0/ 367] loss: 0.277922 [100/ 367] loss: 0.086489 [200/ 367] loss: 0.046113 [300/ 367]

Test Error:

Mean squared error: 0.037214 (baseline: 2.821028)

Epoch 2

loss: 0.033996 [0/ 367] loss: 0.034278 [100/ 367] loss: 0.029303 [200/ 367] loss: 0.027222 [300/ 367]

Test Error:

Mean squared error: 0.027144 (baseline: 2.821028)

Done!

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